



**Barcelona
Supercomputing
Center**

Centro Nacional de Supercomputación



**Earth System Services Group
Earth Sciences Department**



Seasonal climate prediction for the wind energy sector: methods and tools for the development of a climate service

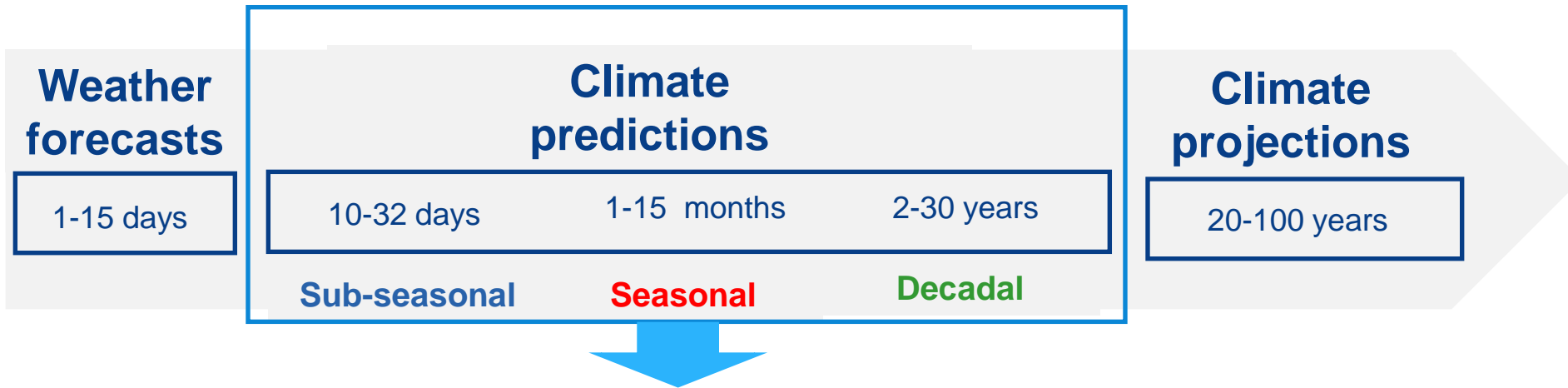
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Tutor: Elsa Mohino

**Jornadas Doctorandos – Universidad Complutense de Madrid
15-17 March 2017**

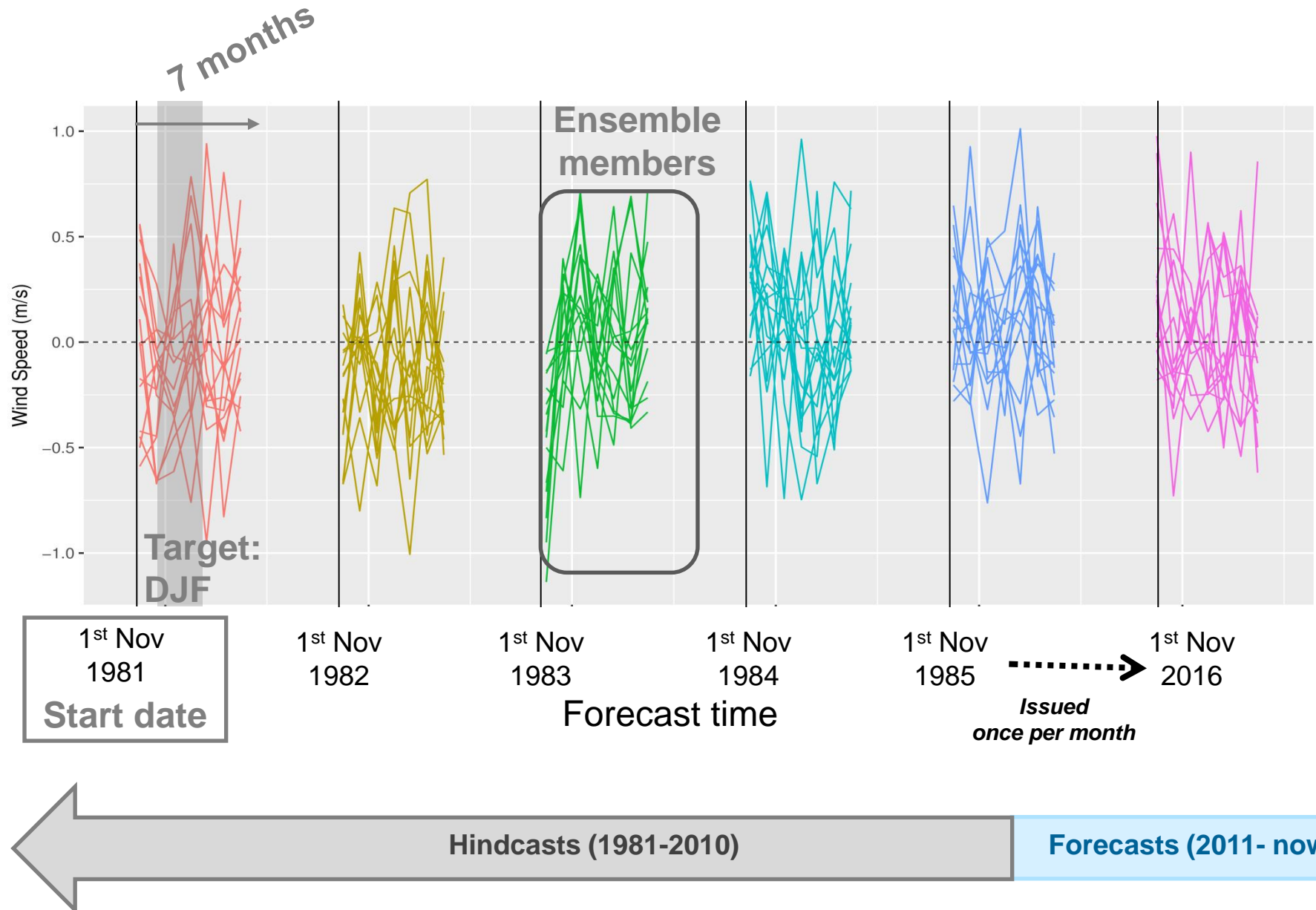
Wind predictability at different time horizons



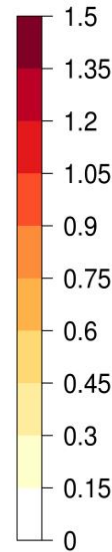
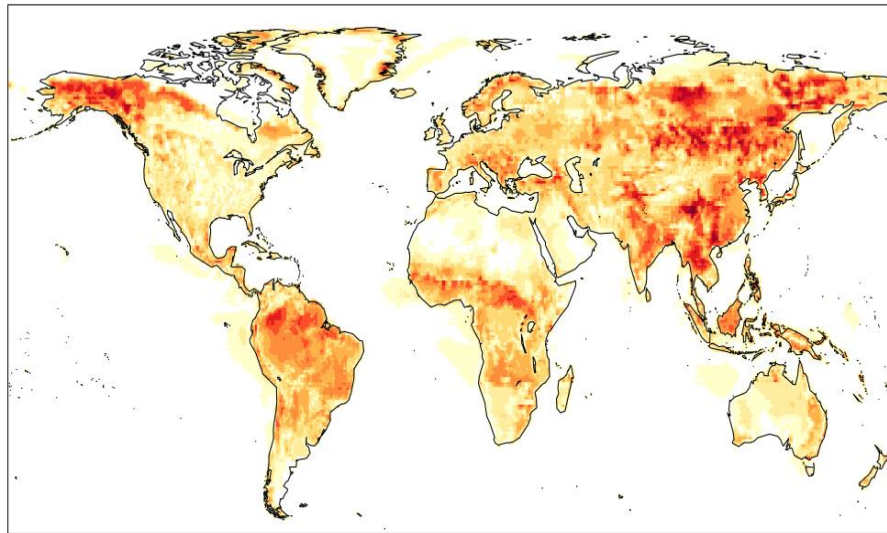
The available seasonal predictions can provide additional value for wind energy applications

- Maintenance works
- Grid management
- Financial issues

ECMWF S4 probabilistic seasonal predictions

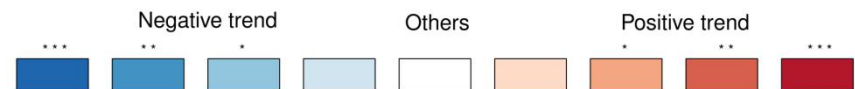
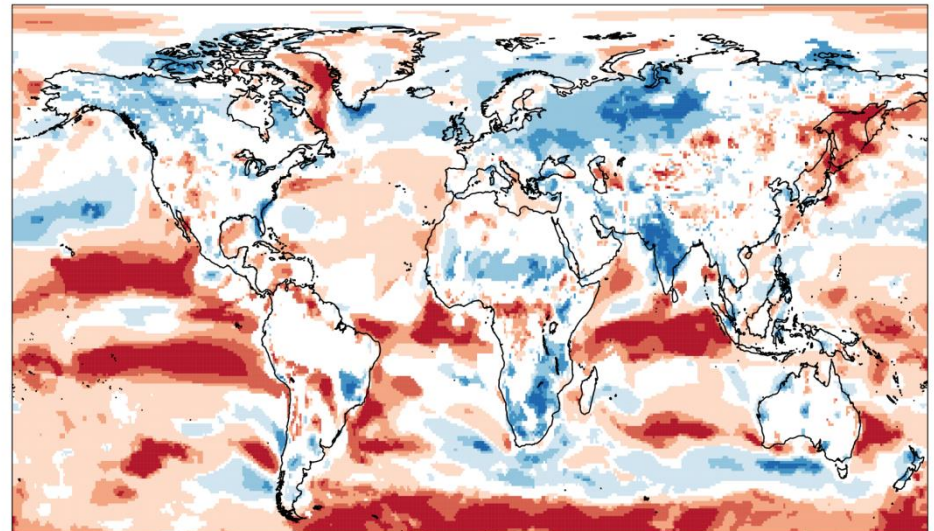


Challenges for the use of climate information: Observational uncertainty

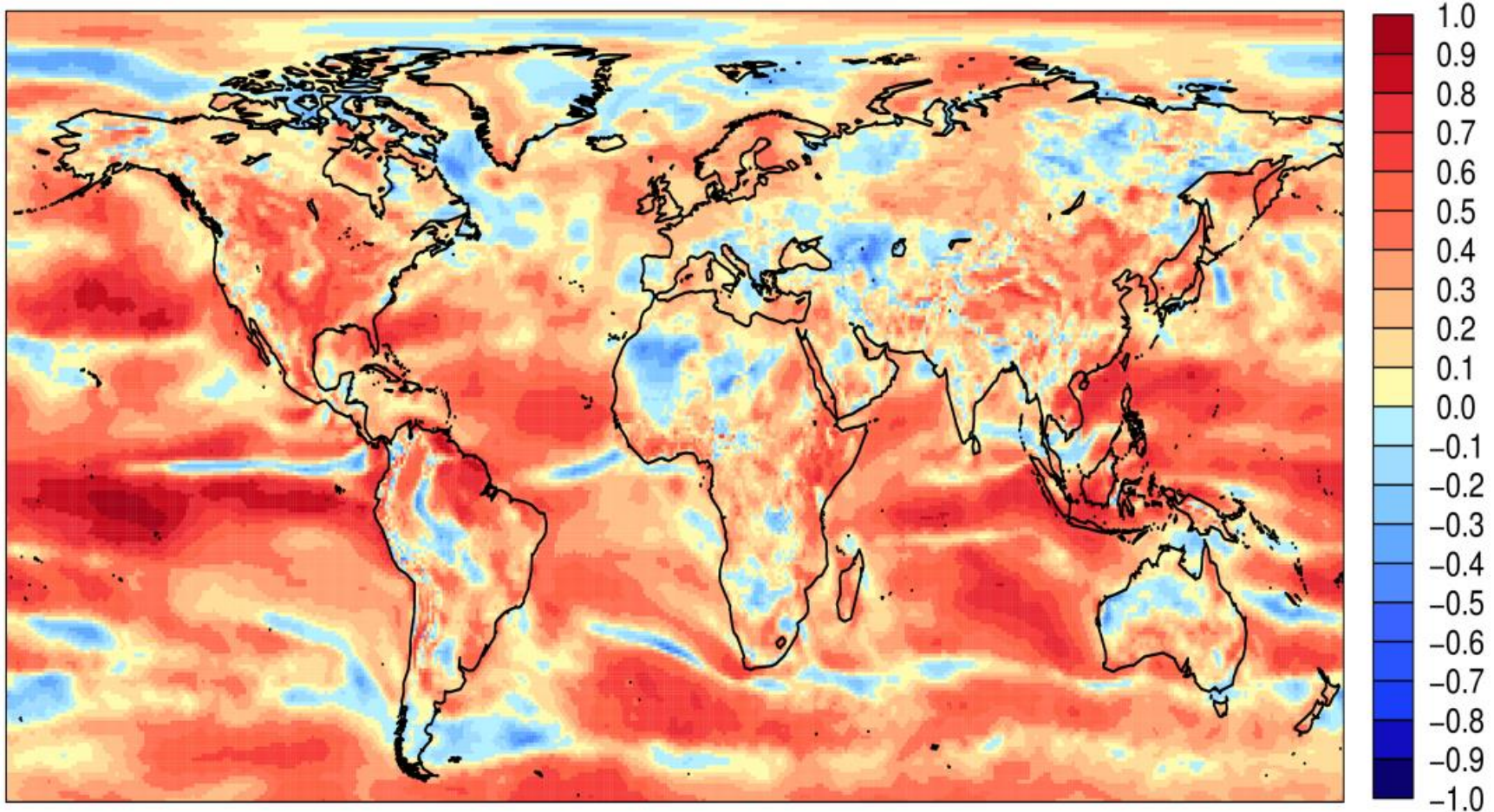


Range of the differences (m/s) between the 10-m wind speed values produced by ERA-Interim, MERRA-2 and JRA-55 reanalyses.

Coherence maps of the 10-m wind speed trends (m/s) in the ERA-Interim, MERRA-2 and JRA-55 reanalyses.

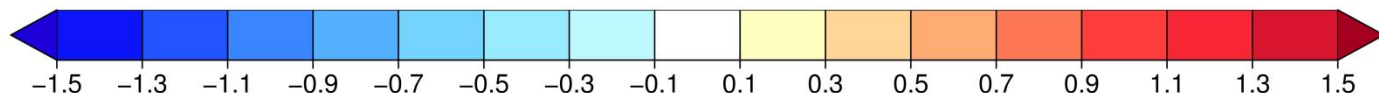
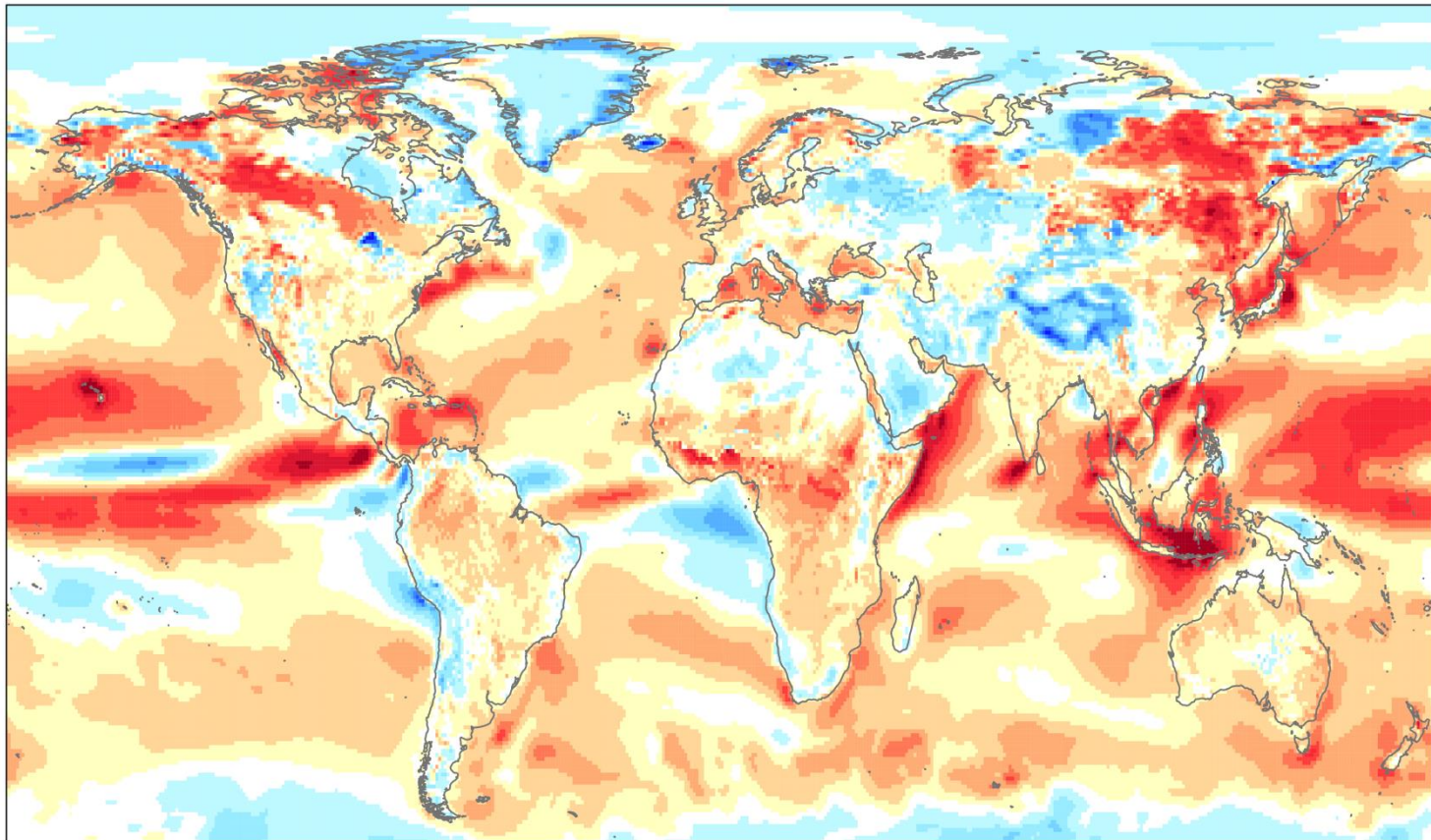


Challenges for the use of climate information: Predictability



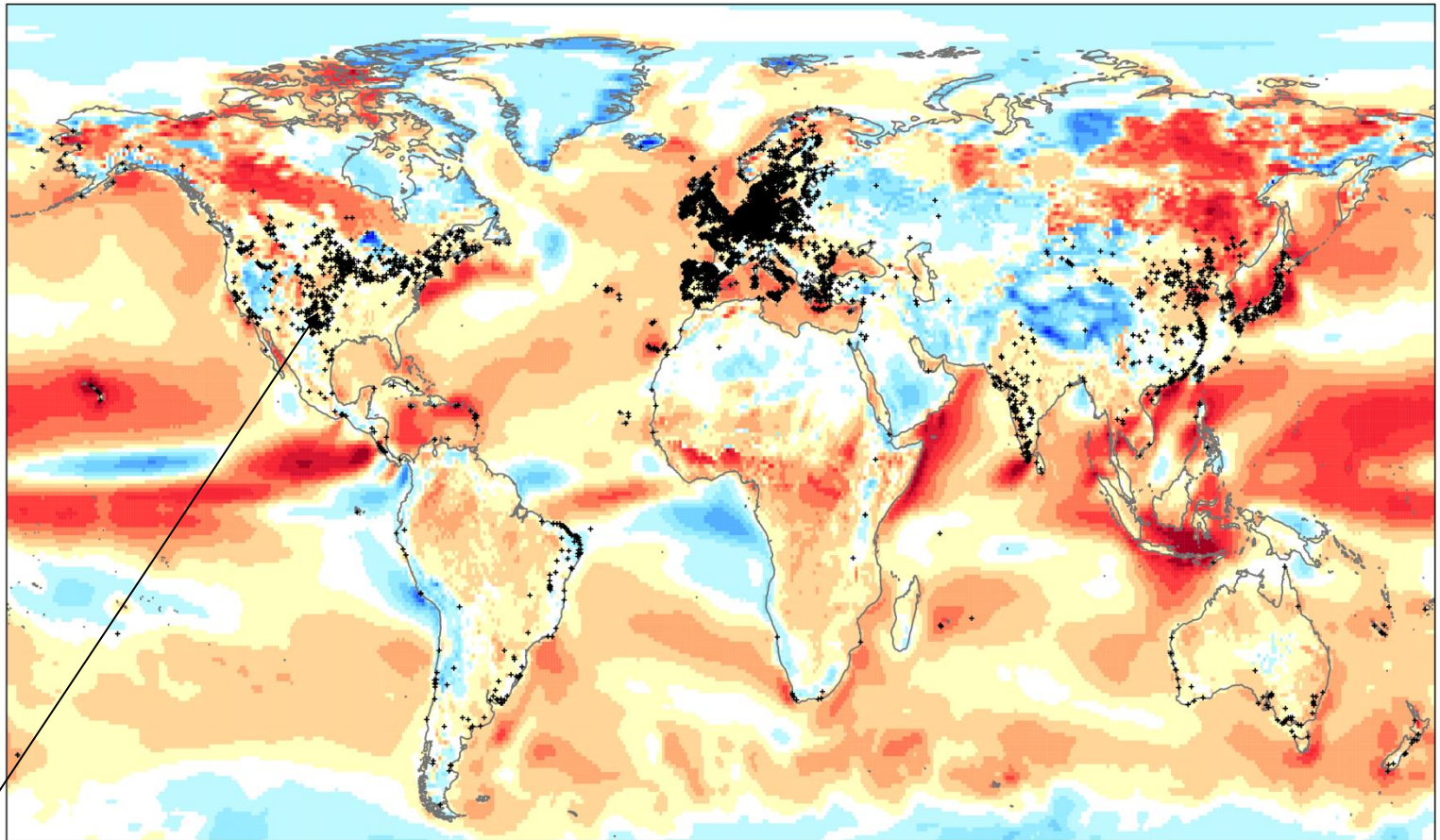
Correlation for 10-m wind speed between the ensemble mean forecasts from ECMWF S4 and ERA-Interim reanalysis in winter

Challenges for the use of climate information: Biases

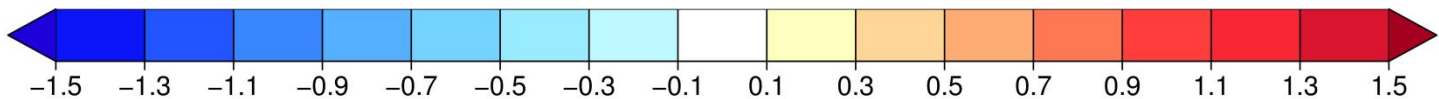


Bias of 10-m wind speed between the ensemble mean forecasts from ECMWF S4 and ERA-Interim reanalysis in winter.

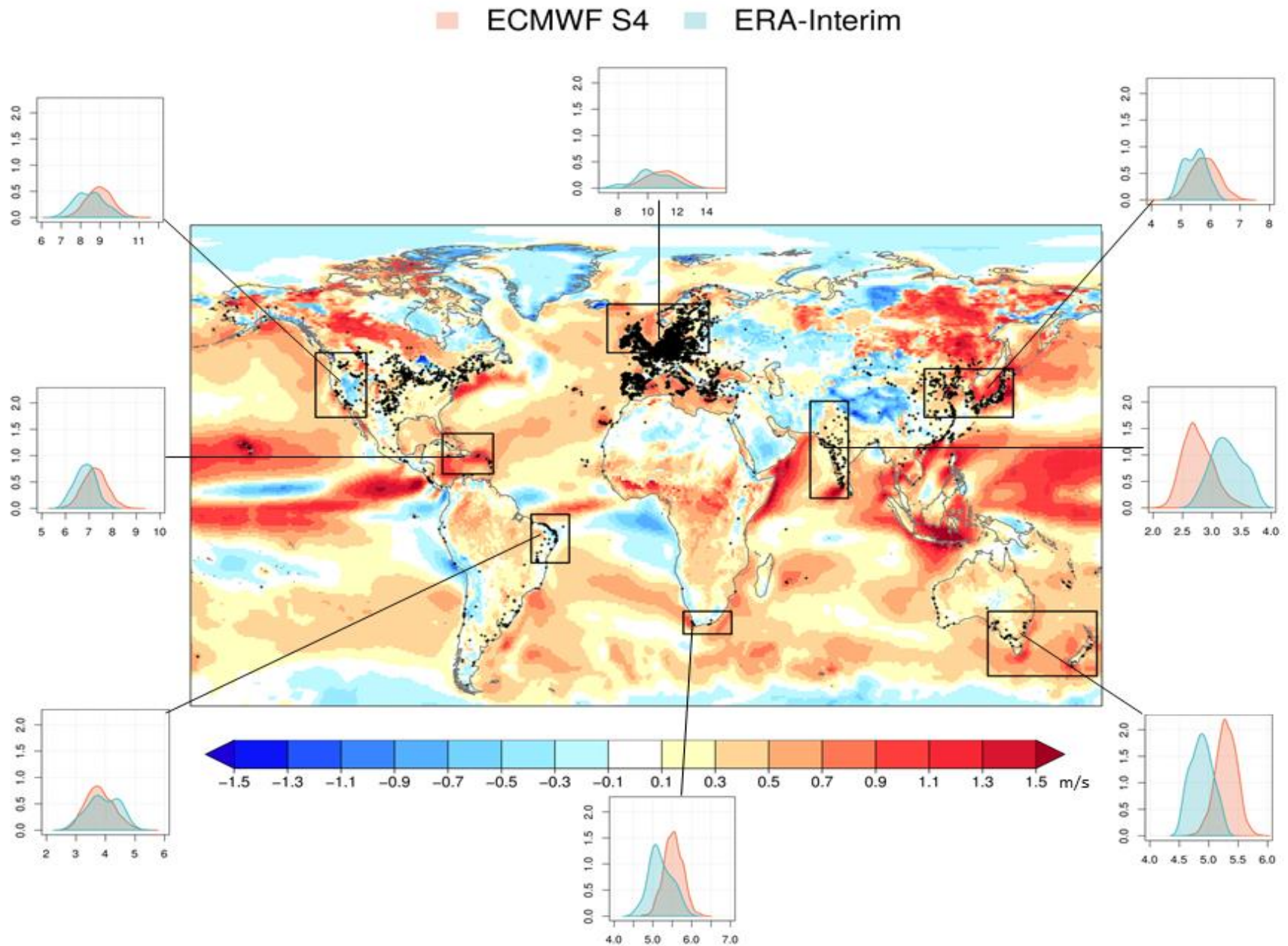
Challenges for the use of climate information: Biases



Installed
wind farms

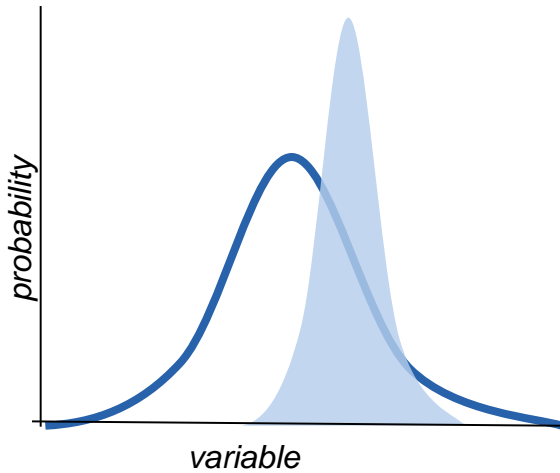


Challenges for the use of climate information: Biases

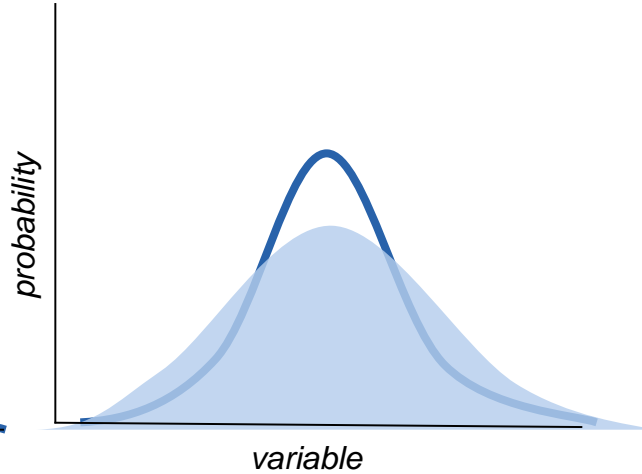


How to develop useful information?

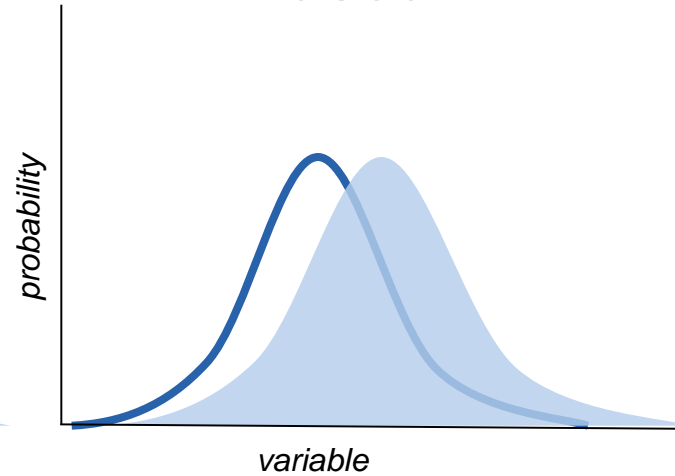
Overconfident



Underconfident

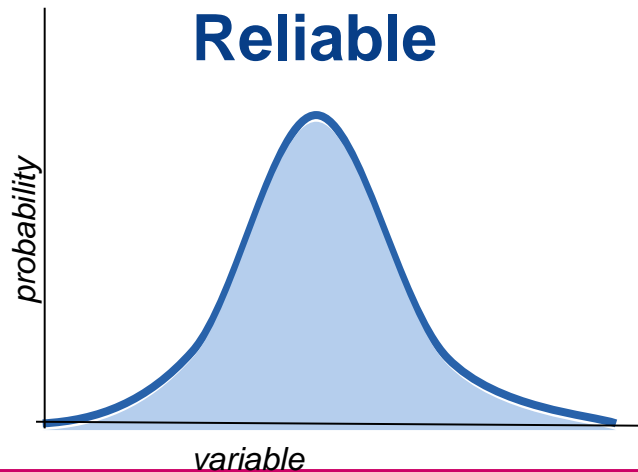


Biased



Bias adjustments

Reliable



Bias-adjustments

Simple bias
correction

Calibration
method

Quantile
mapping

$$y_{j,i} = (x_{ij} - \bar{x}) \frac{\sigma_{ref}}{\sigma_e} - \bar{o}$$

← Variability Observations

← Variability Ensemble

Simple bias correction is based on the assumption that both the reference and forecasted distribution are well approximated by a Gaussian distribution.

Bias-adjustments

Simple bias
correction

Calibration
method

Quantile
mapping

$$y_{j,i} = \alpha x_i + \beta z_{ij}$$

Diagram illustrating the calibration method formula and its components:

- Variability Observations** points to $\alpha = |\rho| \frac{\sigma_o}{\sigma_{em}}$
- Variability Esemble** points to $\beta = \sqrt{1 - \rho^2} \frac{\sigma_o}{\sigma_e}$
- Variability Mean Forecast** points to σ_{em}

Variance inflation:

- Predictions with the same interannual reference as the reference dataset
- Correction of the ensemble spread.

Bias-adjustments

Simple bias
correction

Calibration
method

Quantile
mapping

$$y_{j,i} = ecdf^{ref-1} ecdf^{pred}(x_{ij})$$

Inverse cumulative density
function of the reference
(quantile function)

Cumulative density function of
the predictions

Determines for each forecast to which quantile of the forecast climatology it corresponds, and then maps it to the corresponding quantile of the observational climatology.

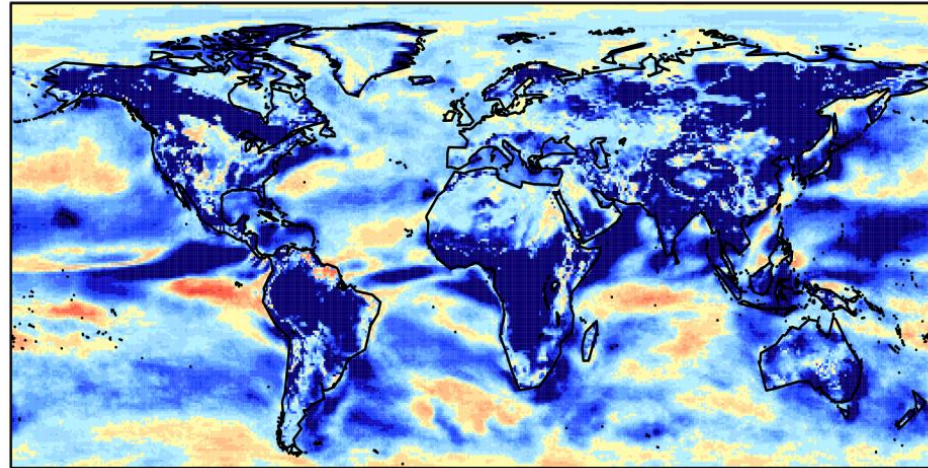
Impact of bias-adjustments on skill

Ranked Probability Skill Score

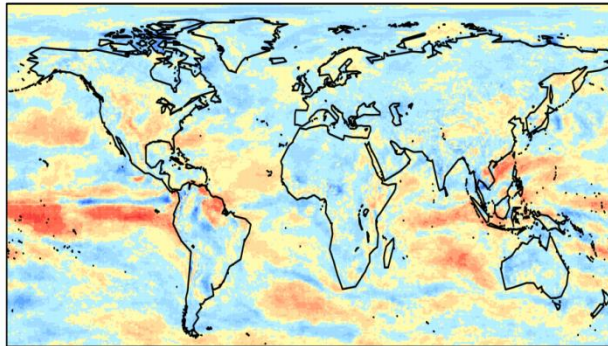
$$RPS = \frac{1}{M-1} \sum_{m=1}^M \left[\left(\sum_{k=1}^m p_k \right) - \left(\sum_{k=1}^m o_k \right) \right]^2$$

$$RPSS = \frac{\overline{RPS} - \overline{RPS}_{reference}}{0 - \overline{RPS}_{reference}} = 1 - \frac{\overline{RPS}}{\overline{RPS}_{reference}}$$

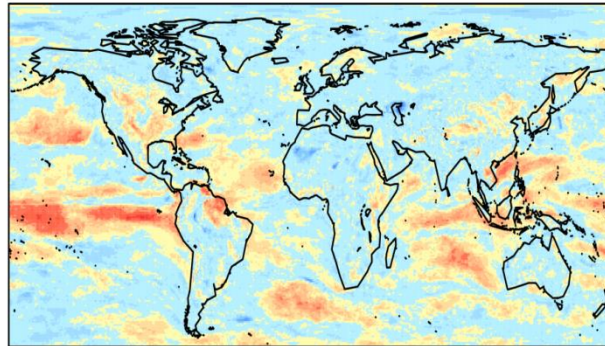
Uncorrected



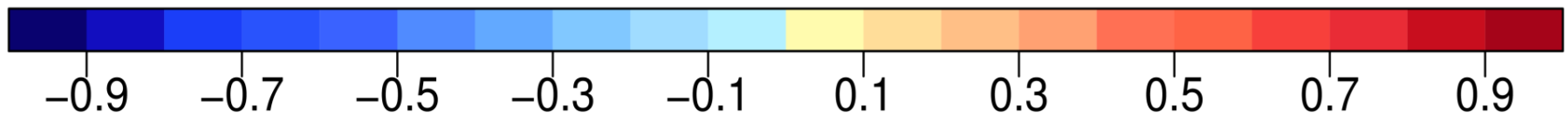
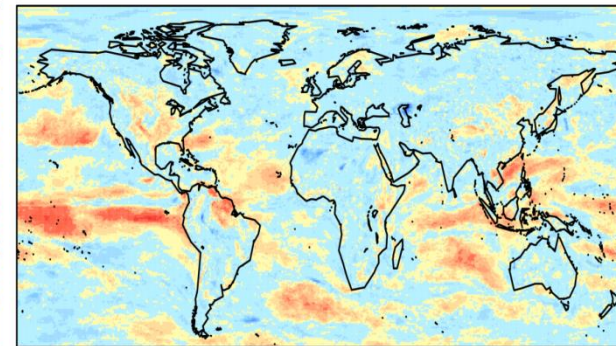
Simple



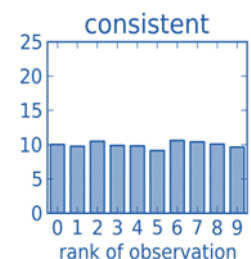
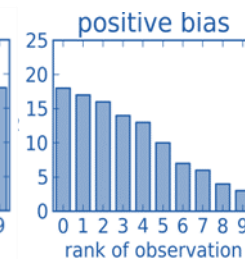
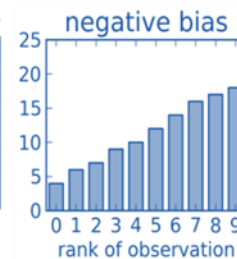
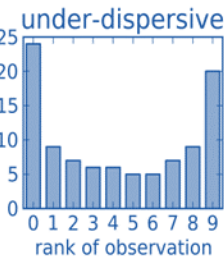
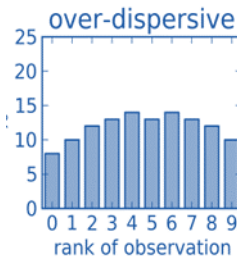
Calibration



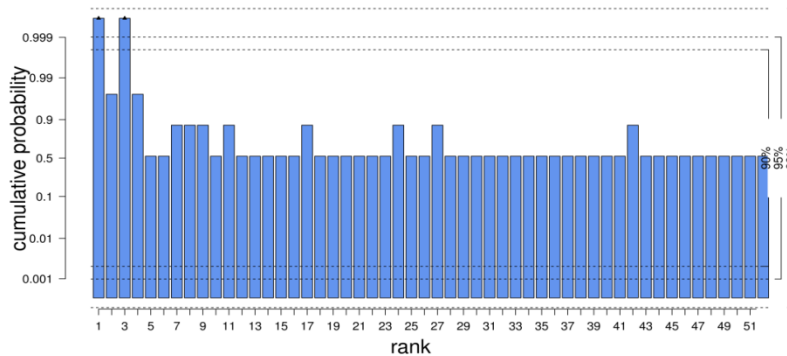
Q-Q mapping



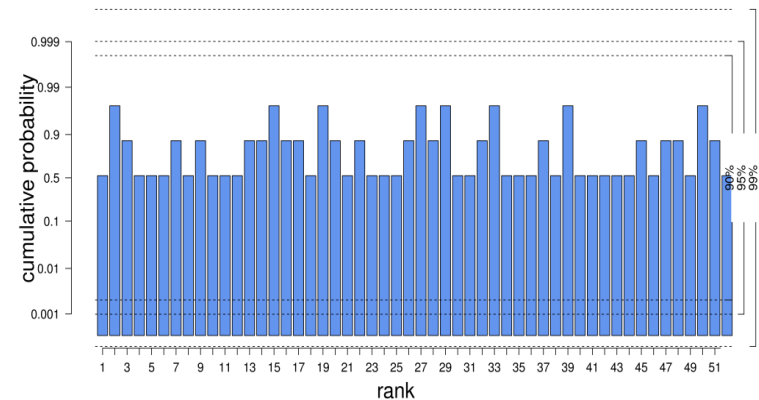
Impact of bias-adjustment on statistical consistency



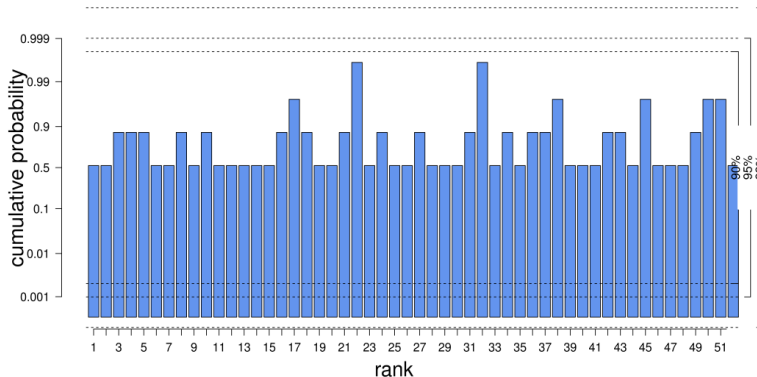
Uncorrected



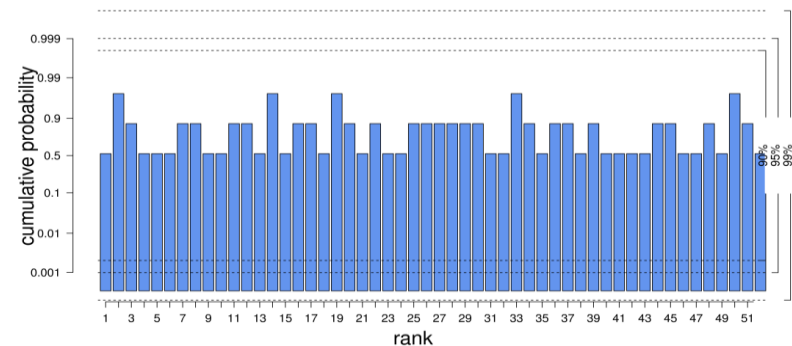
Simple



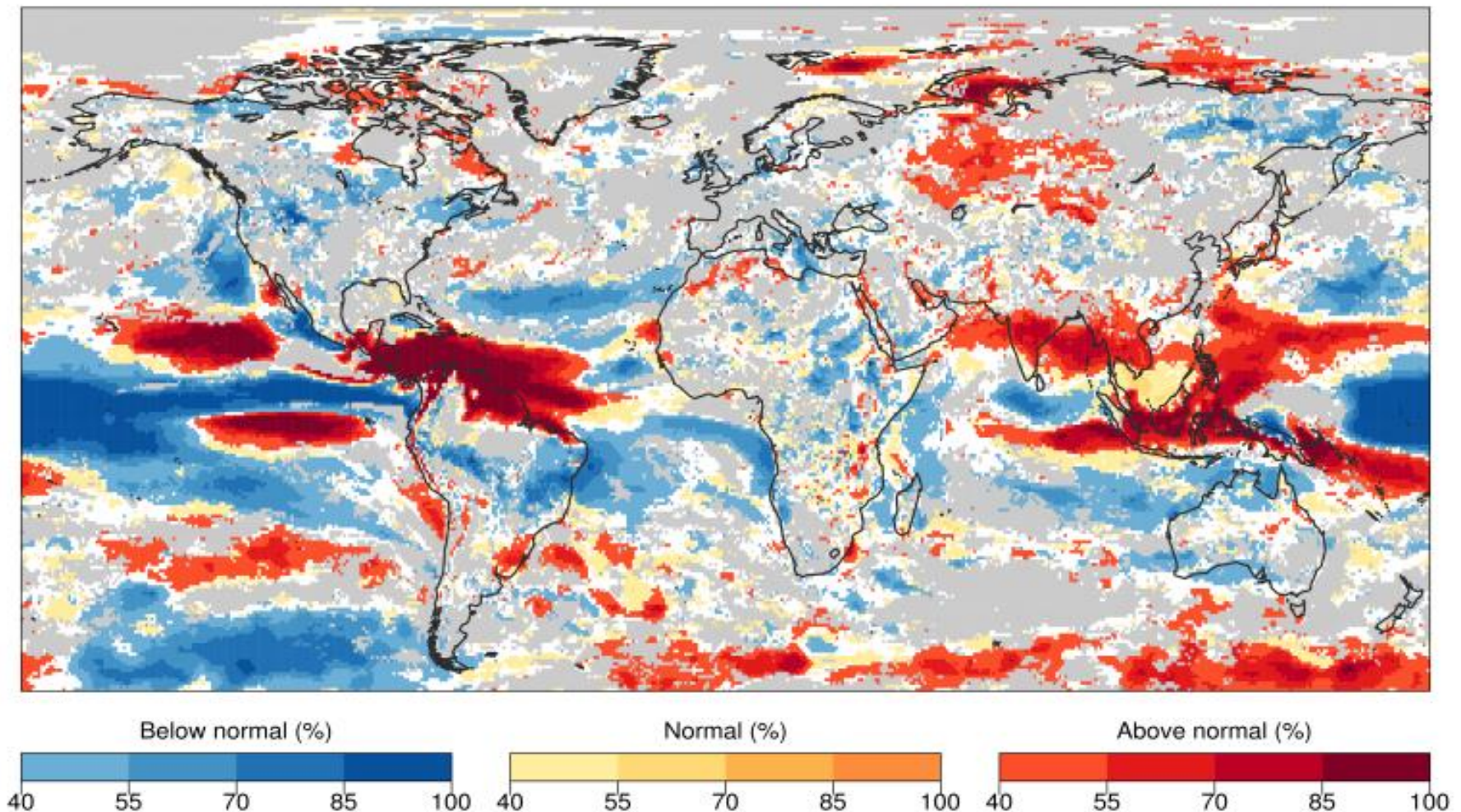
Q-Q mapping



Calibration



Tailored wind speed predictions



ECMWF S4 10-m wind speed seasonal forecast for DJF 2015 initialized the 1st of November. The most likely wind speed category (below-normal, normal or above normal) and its percentage probability to occur is shown. White areas show where the probability is less than 40 % and approximately equal for all three categories. Grey areas show where the climate prediction model doesn't improve the climatology.

Conclusions

- Methodology to develop **useful information for the wind industry**

Torralba V., Doblas-Reyes F.J, MacLeod D., Christel I. and Davis M. (2017) Seasonal climate prediction: a new source of information for the management of the wind energy resources. Journal of applied meteorology and climatolgoy.

- Three methods of bias correction (which are simple enough to be understandable for the users) have been used to produce forecasts with **improved forecast quality**
- Calibration method displays **better reliability** than simple bias correction and quantile mapping, however in terms of skill the three methods produce similar results
- Future work will focus on:
 - Predictions for specific sites
 - Impact model to produce seasonal predictions of wind power capacity factor